



Optimization in Energy Systems Multi-time Step Optimization & Unit Commitment

Lecturer: Dr. Mengshuo Jia





Outline

- Multi-time Step Optimization
 - Motivation, constraints, example
- Unit Commitment Overview
 - Aim, application, challenge
- Deterministic Unit Commitment
 - Integer variables, constraints, objective, example
- Stochastic Unit Commitment
 - Scenario-based approach
- Solutions Method for Unit Commitment
 - Approximation, solutions, validation & reinforcement





Learning Objectives

- After this lecture, you should be able
 - to formulate and solve a multi-time step optimization problem (with solvers) and discuss the advantage of multi-time step versus single time step
 - to formulate the deterministic unit commitment problems
 - to solve the DC-based deterministic unit commitment problem by programming
 - to understand the idea of stochastic unit commitment problems using the scenario-based approach
 - to understand the idea of solving unit commitment problems in reality





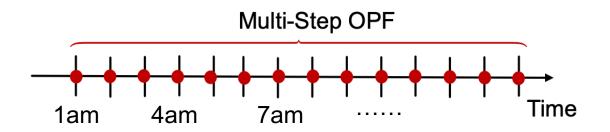
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Multi-time Step Optimization



Single Step OPF

- Motivation: to consider intertemporal constraints
 - o x(k+1) = f[x(k)], e.g., x(k+1) = Ax(k)
 - Storage charging/discharging behavior
 - Ramp limitations for units



Multi-time Step Optimization

Storage constraints

$$\begin{split} E(k+1) &= E(k) + \eta P_{in}(k) - \frac{1}{\eta} P_{out}(k) & \eta \text{, efficiency rate} < 1 \\ 0 &\leq P_{in}(k) \leq P_{in}^{max}, \quad 0 \leq P_{out}(k) \leq P_{out}^{max}, \\ P_{in}(k) \cdot P_{out}(k) &= 0, & \text{Charging or discharging} \\ E^{min} &\leq E(k) \leq E^{max}, \\ E(K) &= E(0) & \text{Restore the energy level} \end{split}$$

Ramp constraints

$$-P_G^{rmax} \le P_G(k+1) - P_G(k) \le P_G^{rmax}$$

A unit cannot significantly change its output immediately



Multi-time Step Optimization

Multi-time DC OPF with storage

$$\min_{P_G} \sum_{k=1}^{K} \sum_{i=1}^{N_G} (a_i P_{G_i}^2(k) + b_i P_{G_i}(k) + c_i)$$

$$s.t. \ P(k) = B\theta(k), \qquad k = 1, ..., K$$

$$E(k+1) = E(k) + \eta P_{in}(k) - \frac{1}{\eta} P_{out}(k), \qquad k = 0, ..., K-1$$

$$E^{min} \leq E(k) \leq E^{max}, \qquad k = 1, ..., K$$

$$P_G^{min} \leq P_G(k) \leq P_G^{max}, \qquad k = 1, ..., K$$

$$0 \leq P_{in}(k) \leq P_{in}^{max}, \qquad 0 \leq P_{out}(k) \leq P_{out}^{max}, \qquad k = 1, ..., K$$

$$P_{in}(k) \cdot P_{out}(k) = 0, \qquad k = 1, ..., K$$

$$-P_{ij}^{max} \leq P_{ij}(k) \leq P_{ij}^{max}, \qquad k = 1, ..., K$$

$$-P_G^{max} \leq P_G(k+1) - P_G(k) \leq P_G^{max}, \qquad k = 0, ..., K-1$$

$$E(K) = E(0)$$



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Unit Commitment – Overview

Aim

Determine an optimal schedule for each generating unit so that

- the demand for electricity is met,
- minimum cost for the system as a whole,

subject to

- ramping constraints,
- network constraints,
- minimum uptime/downtime constraints, etc.

with the decision variables, i.e.,

- generators' outputs
- generators' on/off status





Unit Commitment – Overview

Aim

Determine an optimal schedule for each generating unit so that

- the demand for electricity is met,
- minimum cost for the system as a whole,

subject to

- ramping constraints,
- network constraints,
- minimum uptime/downtime constraints, etc.

with the decision variables, i.e.,

- generators' outputs --- continuous variable
- generators' on/off status --- (integer) binary variable, 1: on, 0: off.





Unit Commitment – Overview

Application

For those with markets:

- Standard tool for clearing electricity markets
- Particularly day-ahead markets

For those without markets:

- Determine the day-ahead commitments and dispatches
- Challenge
 - Mixed-integer nonlinear optimization problem
 - Generally large-scale and nonconvex

Important but challenging





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Deterministic

No uncertainty is considered here

- Variables (Unit j at Time t)
 - Generators' outputs, continuous
 - o $p_i(t)$, power generated
 - Generators' status, binary
 - o $v_i(t)$, on/off status
 - o $y_j(t)$, when starts up
 - o $z_j(t)$, when shuts down





- Binary Decision Variables (Unit j at Time t)
 - $v_i(t)$, on/off status
 - $y_i(t)$, when starts up
 - $z_i(t)$, when shuts down





- Binary Decision Variables (Unit j at Time t)
 - $v_i(t)$, is 1 if unit j is on in time period t, o/w 0
 - $y_j(t)$, is 1 if unit j starts up at the beginning of time period t, o/w 0
 - $z_i(t)$, is 1 if unit j shuts down at the beginning of time period t, o/w 0

Easier to

- ensure that unit j cannot start up and shut down at the same time
- describe the ramping constraints
- describe uptime and downtime constraints
- describe the cost for starting up or shutting down
- 0 ...



- Binary Decision Variables (Unit j at Time t)
 - $v_i(t)$, is 1 if unit j is on in time period t, o/w 0
 - $y_i(t)$, is 1 if unit j starts up at the beginning of time period t, o/w 0
 - $z_j(t)$, is 1 if unit j shuts down at the beginning of time period t, o/w 0
- Logical Coherence between Binary Variables

$$v_j(t-1) - v_j(t) + y_j(t) - z_j(t) = 0$$
 $y_j(t) + z_j(t) \le 1$

- Ensure that $y_j(t)$ and $z_j(t)$ take appropriate values when unit j starts up or shuts down
- E.g., $v_j(0) = 1$ and $v_j(1) = 0$
 - $y_j(1) z_j(1) = -1$, only if $y_j(1) = 0$ and $z_j(1) = 1$ (Hint: binary)





Logical Coherence between Binary Variables

$$v_j(t-1) - v_j(t) + y_j(t) - z_j(t) = 0$$
 $y_j(t) + z_j(t) \le 1$

- Ensure that $y_j(t)$ and $z_j(t)$ take appropriate values when unit j starts up or shuts down
- E.g., for a longer period

	t=0	t=1	t=2	t=3	t=4	t=5
Status (v_j)	0	0	1	1	0	0
Startup (y_j)	0	0	1	0	0	0
Shutdown (z_j)	0	0	0	0	1	0



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- Ramping Constraints
 - In the Continuous OPF Problem:
 - \circ Ramp-up: $p_j(t) p_j(t-1) \leq R_j^{\mathrm{U}}$
 - o Ramp-down: $p_j(t-1) p_j(t) \leq R_j^{\mathrm{D}}$
 - o R_i^U : maximum ramp-up rate of unit j
 - o R_i^D : maximum ramp-down rate of unit j
 - In the UC Problem:
 - \circ Ramp-up: $p_j(t) p_j(t-1) \leq R_j^{\mathrm{U}} \, v_j(t-1) + S_j^{\mathrm{U}} \, y_j(t)$
 - o Ramp-down: $p_j(t-1) p_j(t) \leq R_j^{\mathrm{D}} \, v_j(t) + S_j^{\mathrm{D}} \, z_j(t)$
 - \circ S_i^U : maximum start-up rate of unit j
 - o S_i^D : maximum shut-down rate of unit j



- Ramping Constraints
 - Ramp-up Constraint in UC

$$p_j(t) - p_j(t-1) \le R_j^{U} v_j(t-1) + S_j^{U} y_j(t)$$

- o If $v_j(t-1) = 1$
 - Unit j is on at (t-1), i.e., $y_j(t) = 0$ must hold
 - Because unit j cannot be turned on again
 - Hence: $p_j(t) p_j(t-1) \le R_j^{\mathrm{U}}$
- $\circ \quad \mathsf{lf} \ y_j(t) = 1$
 - Unit j starts up at t, then $v_j(t-1) = 0$ must hold
 - Because unit j cannot be turned on again
 - Hence: $p_j(t) p_j(t-1) \le S_j^U$



- Ramping Constraints
 - Ramp-down Constraint in UC

$$p_j(t-1) - p_j(t) \le R_j^{\mathrm{D}} v_j(t) + S_j^{\mathrm{D}} z_j(t)$$

- $\circ \quad \text{If } v_j(t) = 1$
 - Unit j is on at t, i.e., $z_j(t) = 0$ must hold
 - Because unit j is still on at t
 - Hence: $p_j(t-1) p_j(t) \le R_j^D$
- $\circ \quad \mathsf{lf} \ z_j(t) = 1$
 - Unit j shuts down at t, then $v_j(t) = 0$ must hold
 - Because unit j is already off at t
 - Hence: $p_j(t-1) p_j(t) \leq S_j^D$



- Ramping Constraints
 - Assumption
 - ramp-up rate = start-up rate;
 - ramp-down rate = shut-down rate

$$p_{j}(t) - p_{j}(t-1) \leq R_{j}^{U} v_{j}(t-1) + S_{j}^{U} y_{j}(t)$$

$$p_{j}(t-1) - p_{j}(t) \leq R_{j}^{D} v_{j}(t) + S_{j}^{D} z_{j}(t)$$

$$R_{j}^{U} = S_{j}^{U} \qquad \qquad R_{j}^{D} = S_{j}^{D}$$

$$p_{j}(t) - p_{j}(t-1) \leq R_{j}^{U} [v_{j}(t-1) + y_{j}(t)]$$

$$p_{j}(t-1) - p_{j}(t) \leq R_{j}^{D} [v_{j} + z(t)]$$





- Ramping Constraints
 - Special Case for t = 1

o Up :
$$p_j(t) - p_j(t-1) \le R_j^U [v_j(t-1) + y_j(t)]$$

- o Down: $p_j(t-1) p_j(t) \le R_j^D[v_j + z(t)]$
- $p_j(0) = p_j^{ini}$ and $v_j(0) = v_j^{ini}$ are parameters i/o variables
- o To distinguish:

$$\begin{array}{lll} & - & p_{j}(t) - p_{j}^{ini} \leq R_{j}^{U} \big[v_{j}^{ini} + y_{j}(t) \big], & \forall j, & t = 1 \\ & - & p_{j}(t) - p_{j}(t-1) \leq R_{i}^{U} \big[v_{j}(t-1) + y_{j}(t) \big], & \forall j, & \forall t > 1 \end{array}$$

$$\begin{array}{lll} & - & p_{j}^{ini} - p_{j}(t) \leq R_{j}^{D} \big[v_{j} + z(t) \big], & \forall j, \ t = 1 \\ & - & p_{j}(t-1) - p_{j}(t) \leq R_{j}^{D} \big[v_{j} + z(t) \big], & \forall j, \ \forall t > 1 \end{array}$$





- Uptime/downtime Constraints
 - Description: a unit cannot be turned on or off arbitrarily
 - o If unit j starts up at time period k
 - Then it must stay "on" for T_j^U time periods (including the time period k itself) => uptime constraint
 - \circ If unit j shuts down at time period k
 - then it must stay "off" for T_j^D time periods (including the time period k itself) => downtime constraint





Uptime constraint

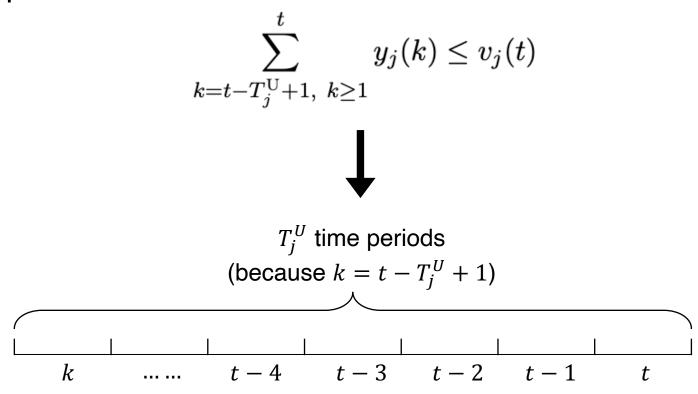
$$\sum_{k=t-T_j^{\mathrm{U}}+1,\ k\geq 1}^t y_j(k) \leq v_j(t)$$

- If unit j starts up at time period k
 - O Then it must stay "on" for T_j^U time periods (including the time period k itself)
- o How it works?





Uptime constraint





Uptime constraint

$$\sum_{k=t-T_j^{\mathrm{U}}+1,\ k\geq 1}^t y_j(k) \leq v_j(t)$$

$$\downarrow$$

$$\Sigma_j^U \leq v_j(t)$$

$$k$$
 $t-4$ $t-3$ $t-2$ $t-1$ t
 $y_{j}=1$ $y_{j}=0$ $y_{j}=0$ $y_{j}=0$ $y_{j}=0$ $y_{j}=0$ $y_{j}=0$
 $v_{j}=1$ $v_{j}=1$ $v_{j}=1$ $v_{j}=1$ $v_{j}=1$ $v_{j}=1$

o If unit j starts up at time period k, it must stay "on" for T_j^U time periods (including the time period k itself)

Unit j starts up at time period k

$$\circ \Rightarrow y_j(k) = 1$$

$$\circ \Rightarrow v_i(k) = 1$$

$$\circ \Rightarrow \Sigma_i^U \geq 1$$

$$\circ$$
 $\Sigma_i^U > 1 \text{ or } \Sigma_i^U = 1?$

$$\circ \quad \mathsf{lf} \ \Sigma_i^U > 1$$

o Maximal
$$v_i(t)$$
 is 1

o
$$\Sigma_i^U > 1 = v_i(t)_{max}$$
, contradictory!

$$\circ \Rightarrow \Sigma_i^U = 1$$

$$\circ \quad \Sigma_j^U = 1$$

$$\circ \quad \Rightarrow y_j(t), y_j(t-1), \dots, y_j(k+1) = 0$$

$$\circ \quad \Sigma_i^U \leq v_j(t) \text{ and } \Sigma_i^U = 1$$

$$\circ \Rightarrow v_j(t) = 1$$

o If
$$v_j(m) = 0$$
, $m \in [k+1, t-1]$?

$$o \Rightarrow y_i(n) = 1, \exists n \in [m+1, t]$$

o But
$$y_i(t), y_i(t-1), ..., y_i(k+1) = 0$$

$$o \Rightarrow v_i(t-1), \dots, v_i(k+1) = 1$$



Downtime constraint

$$v_j(t) + \sum_{k=t-T_j^{D}+1, k \ge 1}^{t} z_j(k) \le 1$$

- o If unit j shuts down at time period k
 - o then it must stay "off" for T_j^D time periods (including the time period k itself)
- O How it works?

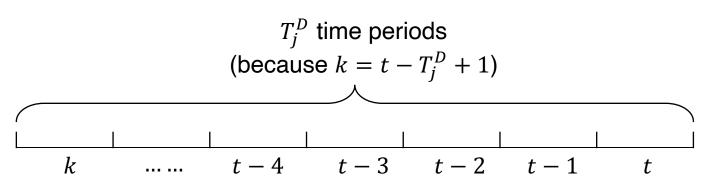




Downtime constraint

$$v_j(t) + \sum_{k=t-T_j^{D}+1, k \ge 1}^{t} z_j(k) \le 1$$







Downtime constraint

$$egin{aligned} v_j(t) + \sum_{k=t-T_j^{\mathrm{D}}+1, \ k \geq 1}^t z_j(k) \leq 1 \ & iggrup v_j(t) + \Sigma_j^D \leq 1 \end{aligned}$$

o If unit j shuts down at time period k, it must stay "off" for T_j^D time periods (including the time period k itself)

$$\circ$$
 Unit *j* shuts down at time period k

$$\circ \Rightarrow z_i(k) = 1$$

$$\circ \Rightarrow v_i(k) = 0$$

$$\circ \Rightarrow \Sigma_i^D \geq 1$$

$$\circ \quad \Sigma_i^D > 1 \text{ or } \Sigma_i^D = 1?$$

$$\circ \quad \mathsf{lf} \ \Sigma_i^D > 1$$

o Minimal
$$v_i(t)$$
 is 0

o
$$v_i(t)_{min} + \Sigma_i^D > 1$$
, contradictory!

$$\circ \Rightarrow \Sigma_i^D = 1$$

$$\circ \quad \Sigma_j^D = 1$$

$$\circ$$
 $z_i(k) = 1$ and $\Sigma_i^D = 1$

$$\circ \quad v_j(t) + \Sigma_j^D \le 1 \text{ and } \Sigma_j^D = 1$$

$$\circ \Rightarrow v_i(t) = 0$$

$$o If v_j(m) = 1, m \in [k+1, t-1]?$$

$$\circ$$
 $\Rightarrow z_i(n) = 1, \exists n \in [m+1, t]$

o But
$$z_i(t), z_i(t-1), ..., z_i(k+1) = 0!$$

$$o \Rightarrow v_i(t-1), \dots, v_i(k+1) = 0$$



- Generation Constraints
 - In the Continuous OPF Problem:
 - Upper limit: $p_j(t) \le \overline{p}_j$
 - Lower limit: $p_j \le p_j(t)$
 - In the UC Problem:
 - Upper limit: $p_j(t) \le \overline{p}_i v_j(t)$
 - Lower limit: $\underline{p}_j v_j(t) \le p_j(t)$
 - Still linear





- Objective Function
 - In the Continuous OPF Problem:

$$\min_{\Xi} \quad \sum_{t \in T} \sum_{j \in J} \left[c_j(p_j(t)) \right]$$

- o $c_i(p_i(t))$ represents the generation cost of unit j at time t
- In the UC Problem:

$$\min_{\Xi} \sum_{t \in T} \sum_{j \in J} \left(c_j(p_j(t)) + c_j^{\mathrm{U}} y_j(t) + c_j^D z_j(t) \right)$$

- c_j^U is a constant cost for starting up unit j
- \circ c_i^D is a constant cost for shutting down unit j





Example: DC-based UC



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Stochastic Unit Commitment

- Uncertainty
 - Stochastic sources
 - Most notably wind and solar generation
 - Demand, yet can be assumed to be deterministic (as in this lecture)
- Modelling
 - Stochastic (Scenario-based) optimization
 - Chance-constrained optimization
 - Robust optimization
 - Distributionally robust optimization





- Uncertainty
 - Stochastic sources
 - Most notably wind and solar generation
 - Demand, yet can be assumed to be deterministic (as in this lecture)
- Modelling
 - Stochastic (Scenario-based) optimization for UC problems
 - Uncertainties from wind power generators
 - Chance-constrained optimization
 - Robust optimization
 - Distributionally robust optimization



- Stochastic UC Formulation
 - Two-stage stochastic programming problem
 - First stage
 - Make decisions in advance, e.g., day-ahead UC
 - Uncertainties have not been realized
 - Assume renewable generators are dispatchable
 - Like conventional generators
 - Similar to a deterministic UC problem
 - Difference: introduce reserve variables $r_i^+(t)$, $r_i^-(t)$





- Stochastic UC Formulation
 - Two-stage stochastic programming problem
 - Second stage
 - To make decisions in "real-time"
 - Uncertainties => realizations, e.g., wind power realizations
 - Adjust the first stage decisions after observing the realizations
 - Reserve deployment variables $p_{i\omega}^+(t)$, $p_{i\omega}^-(t)$
 - Wind power spillage variables $w_{j\omega}^{spil}(t)$
 - Load shedding variables $l_{\omega}^{shed}(t)$





- Stochastic UC Formulation
 - Two-stage stochastic programming problem
 - Second stage
 - To make decisions in "real-time"
 - Uncertainties => realizations, e.g., wind power realizations
 - Adjust the first stage decisions after observing the realizations
 - In fact, no realizations have been seen => one day ahead
 - Assume a set of realizations of wind power generations
 - A realization => a scenario
 - Do adjustment for each scenario





- Stochastic UC Formulation
 - Two-stage stochastic programming problem
 - Combine the first and second stage together
 - The decisions made in the first stage can satisfy all possible scenarios in the second stage
 - Conventional generators' status
 - Conventional generators' output
 - Conventional generators' reserves provided





- First Stage
 - Variables
 - $w_i(t)$, power generated by wind power unit i at time t
 - $\circ r_i^+(t)$, upward reserve provided by conventional unit i at time t
 - \circ $r_i^-(t)$, downward reserve provided by conventional unit i at time t
 - o Reserve:
 - Amount of additional capacity available beyond the actual power output
 - Typically used to provide a buffer
 - In cases of, e.g., unexpected increases/decreases in renewable generations





- First Stage
 - Variables
 - $w_i(t)$, power generated by wind power unit i at time t
 - o $r_i^+(t)$, upward reserve provided by conventional unit i at time t
 - \circ $r_i^-(t)$, downward reserve provided by conventional unit i at time t
 - o $p_i(t), v_i(t), y_i(t), z_i(t)$
 - Active power generation of unit i at time t
 - Commitment variables of unit i at time t





First Stage

Main Constraints

o
$$y_i(t) - z_i(t) = v_i(t) - v_i(t-1), \forall i, t$$

$$y_i(t) + z_i(t) \le 1, \forall i, t$$

$$o p_i(t) + r_i^+(t) \le \overline{p}_i v_i(t), \forall i, t$$

$$o \underline{p_i}v_i(t) \leq p_i(t) - r_i^-(t), \forall i, t$$

Power Balance

Commitment Status

Commitment Status

Generation upper limit

Generation lower limit





- First Stage
 - Other Constraints

$$v_i(t), y_i(t), z_i(t) \in \{0, 1\}, \forall i, t, t$$

Binary restriction

$$0 \le w_j(t) \le \overline{w}_j, \, \forall j, t$$

Wind power bounds

- Omit uptime/downtime, ramping, network constraints
- Objective

- Minimize the overall cost of
 - Generation, start up, shut down
 - Providing upward reserve and downward reserve





- Second Stage
 - Scenario
 - \circ Use index ω to denote scenario ω
 - Ouse $w_{j\omega}^*(t)$ to denote the realization of the output of wind power unit j in scenario ω at time t --- parameter
 - Introduce new Variables for adjustments
 - o $p_{i\omega}^+(t)$, $p_{i\omega}^-(t)$, upward/downward reserve deployments from conventional unit i at time t under scenario ω
 - o $w_{j\omega}^{spil}(t)$, power spillage from wind power unit j at time t under scenario ω
 - o $l_{\omega}^{shed}(t)$, overall load shedding at time t under scenario ω



Second Stage

- Why there is a need for adjustment?
 - \circ Scheduled output from the first stage: $w_i(t)$
 - O Realized output from scenario ω: $w_{i\omega}^*(t)$
 - O Deviation: $\sum_{j} \left[w_{j\omega}^{*}(t) w_{j}(t) \right]$ --- aim to eliminate the deviation
- Adjustment Method
 - $\sum_{i} [p_{i\omega}^{+}(t) p_{i\omega}^{-}(t)]$, the regulation provided by all conv. units
 - $\sum_{i} \left[-w_{i\omega}^{spil}(t) \right]$, the spillage provided by all wind power units
 - $l_{\omega}^{shed}(t)$, the load shedding provided by the overall load
- Aim



Second Stage

Constraints

$$\circ \quad \sum_{j} \left[w_{j\omega}^{*}(t) - w_{j}(t) - w_{j\omega}^{spil}(t) \right] + \sum_{i} \left[p_{i\omega}^{+}(t) - p_{i\omega}^{-}(t) \right] + l_{\omega}^{shed}(t) = 0 \; , \; \forall t, \omega$$

$$0 \le p_{i\omega}^+(t) \le r_i^+(t)$$
, $\forall i, t, \omega$. Recall $r_i^+(t)$ is from the first stage

o
$$0 \le p_{i\omega}^-(t) \le r_i^-(t)$$
, $\forall i, t, \omega$. Recall $r_i^-(t)$ is from the first stage

$$0 \le l_{\omega}^{shed}(t) \le L_t, \ \forall t, \omega$$

$$0 \le w_{j\omega}^{spil}(t) \le w_{j\omega}^*(t), \ \forall j, t, \omega$$



- Second Stage
 - Objective

$$\circ \mathbb{E}\left\{\sum_{t}\left(\sum_{i}\left[c_{i}\left(p_{i\omega}^{+}(t)-p_{i\omega}^{-}(t)\right)\right]+c^{shed}l_{\omega}^{shed}(t)\right)\mid\forall\omega\right\}$$

- Minimize the expectation of the regulation cost of
 - Reserve deployments, load shedding, for each scenario
 - Wind power is for free, i.e., spillage has no cost
- \circ Assume π_{ω} is the probability of scenario ω --- known
- $\circ \quad \mathbb{E}\{\cdot \mid \forall \omega\} = \sum_{\omega} \pi_{\omega} \ \sum_{t} \left(\sum_{i} \left[c_{i} \left(p_{i\omega}^{+}(t) p_{i\omega}^{-}(t) \right) \right] + c^{shed} l_{\omega}^{shed}(t) \right)$





- Stochastic UC Problem
 - Objective
 - Objective of the 1st stage + Objective of the 2nd stage
 - Constraints
 - Union of the constraints in the 1st and 2nd stages
 - Variables
 - Union of the variables in the 1st and 2nd stages
 - However, only some variables in the 1st stage will be outputted
 - Conventional generators' status
 - Conventional generators' output
 - Conventional generators' reserves provided





- Drawbacks
 - Assume explicit knowledge of the probability distribution of the uncertainties.
 - Estimated empirically
 - Data-driven
 - Simulation-driven
 - Impact the quality of the results
 - As the number of scenarios increases, the optimization problem become larger and more challenging
 - Large scale => Solvers may fail even for mixed integer linear programming



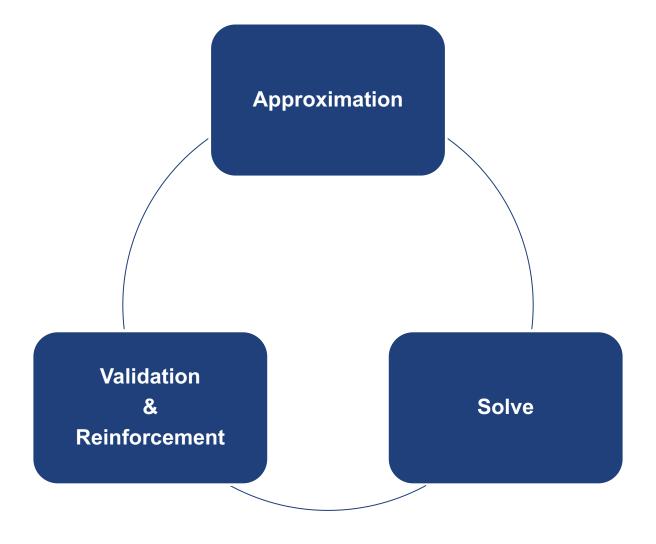


Outline

- Multi-time Step Optimization
 - Motivation, constraints, example
- Unit Commitment Overview
 - Aim, application, challenge
- Deterministic Unit Commitment
 - Integer variables, constraints, objective, example
- Stochastic Unit Commitment
 - Scenario-based approach
- Solutions Method for Unit Commitment
 - Approximation, solutions, validation & reinforcement











- Approximation
 - AC-based UC
 - Mixed integer nonlinear programming (MINLP) --- challenging
 - DC approximation => DC-based UC
 - Mixed integer linear programming (MILP) --- Solvable
 - However, large scale => still challenging
 - Keep the number of scenarios in stochastic UC small
 - Research for scenario and/or scale reduction
 - Most common approximation method
 - Question => how to solve an MILP problem?





- Solve
 - State-of-the-art Solvers
 - E.g., Gurobi, CPLEX, AMPL, GAMS => Enough for this lecture
 - The common method to solve UC on a day-ahead basis
 - Algorithms Inside
 - General principles: divide-and-conquer
 - Recursively partitions the feasible region
 - => Find integer (binary) solutions
 - Typical algorithm: branch-and-bound (BnB)
 - Most (mixed) integer optimization solvers use some form of branch-and-bound algorithm



- Solve
 - Branch-and-bound algorithm: basic idea
 - Relaxation
 - General MILP: remove integer restrictions, i.e., $x \in \mathbb{R}$
 - 0/1 MILP: replace 0/1 restrictions with $0 \le x \le 1$
 - Branch/divide/add constraints with relaxation
 - General MILP: $x \le N$ and $x \ge N + 1 \Rightarrow$ two subproblems
 - E.g., Min f(x), s.t. $g(x) \le 0$. If $x^* = 5.4$, then
 - Add $x \le 5$ to $g(x) \le 0$, resolve, see if $x^* = 5$, or 4, ...
 - Add $x \ge 6$ to $g(x) \le 0$, resolve, see if $x^* = 6$, or 7, ...
 - 0/1 MILP: x = 0 and $x = 1 \Rightarrow$ two subproblems





- Validation & Reinforcement
 - DC-based UC Result
 - Optimal solution for the approximated problem
 - May not even feasible for the original AC-based problem
 - Check & Add Constraints
 - Input the DC-based UC solutions into AC power flow equations
 - Violations of transmission flows and/or bus voltages?
 - If so, add additional constraints to the DC-based UC
 - Re-solve & re-validation & re-reinforcement





Summary

- Multi-time Step Optimization
 - Motivation, ramping constraints, storage constraints, receding horizon
- Unit Commitment Overview
 - Aim, application, challenge
- Deterministic Unit Commitment
 - Integer variables, logic constraints, ramping constraints, uptime/downtime constraints, generation Constraints, objectives
- Stochastic Unit Commitment
 - Scenarios, additional variables and constraints, two-stage formulations
- Solutions Method for Unit Commitment
 - Approximation, solutions (BnB), validation & reinforcement

